



Advanced Information Systems Technologies (AIST)

Machine Learning in NASA's Earth Science Division



Michael Little
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Outline of Talk

- Opportunities
- What's it Going to Take
- Analytic Centers as a Framework
- Concept of Operations
- Some Experiments
- Strategy



Technology Opportunities

- Generating data and numeric output
 - Constellations of SmallSats with in situ, airborne platforms
 - Closer interaction with physical and statistical models
 - Autonomy
- Computing Capabilities
 - Public Cloud Computing
 - Quantum Computing
 - Cognitive Computing
- Exploiting Data
 - Machine Learning
 - Workflow tools, such as Notebooks
 - Integration of data into a unified picture of a natural phenomenon or physical process
 - Virtual Reality



Machine Learning Hopes

- A more robust **understanding** of a natural phenomenon or physical process
 - Relying on conclusions based on 3,000 examples, not 3
 - Digest massive volumes of observational data
 - High volume output from high-resolution ES models
 - Conventional techniques drown ES researchers
- **Prediction** of future state for complex systems from observational data without a thorough understanding of complex or poorly understood underlying physics
- **Coordination** of elements of a sensor web to target transient and transitional phenomena
 - Leverage emergence of SmallSats
 - Autonomy
 - Detection of interesting features and re-tasking to observe them
- Fast, comparative **OSSEs**
 - What is the relative value of science for different Observing Systems
- **Assimilation** models which do better at washing out the experimental error while preserving features and physical anomalies
- A way to **clean up observational data** to make it easier to analyze



Lessons Learned (so far)

- ML and advanced tools are not trusted by Science Community
 - Not well understood, including which ones to use for what
 - Difficult to use
 - Don't know who to trust
- Current crop of tools are hard to use
- Data is hard to coerce into an analyzable format
 - Finding usable data is hard (current catalogs offer limited help)
 - Large volumes of data are hard to process
 - Diverse formats, uncertainties, sources are hard to integrate
- Non-Earth Science Community is far ahead in using them
 - Financial Markets
 - Computer Security (and other security)
 - Commercial Space



What's it Going to Take?

- **Community Acceptance** in NASA Earth Science (ES)
 - NASA Earth Scientists must understand the processes & algorithms
 - “In the field of observation, fortune favors the prepared mind.” – Pasteur
 - Collaboration among Earth Scientists, Computer Scientists and Technologists
 - Demonstration of value to the science communities
 - New Science
 - Faster Time to Science, retaining quality and understanding
 - Results backed up by confirmation with legacy techniques
 - Uncertainty Quantification
 - Engagement with NASA ES by experienced ML practitioners
- **Improved Data Usability**
 - Rapid use of the numeric results without time-delays for grooming
- **Easy to use** tools and credible results
 - Need a Framework to integrate tools, data and computation resources

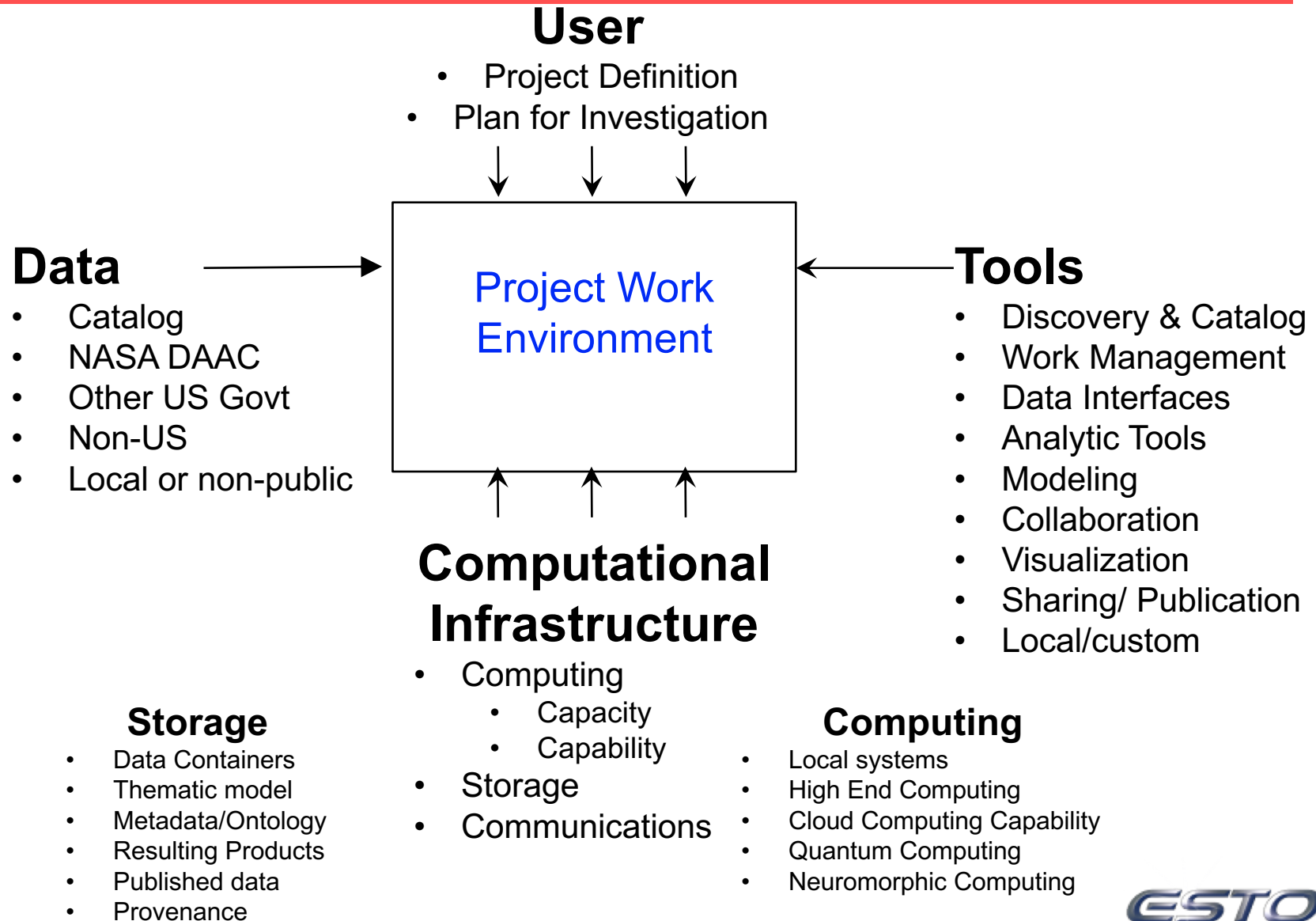


A Way to Focus Technology Development

- An environment for conducting a **Science investigation**
 - Enables the confluence of resources for that investigation
 - Tailored to the individual study topic
- Harmonizes **data, tools and computational resources** to permit the research community to focus on the investigation
 - Reduce the data preparation time to something tolerable
 - Catalog of optional resources (think HomeDepot shopping)
 - Semantic-enabled catalog of resources (think Yelp) with help
 - Relevant publications
 - Provide established training data sets of varying resolution
 - Provide effective project confidentiality, integrity and availability
 - Single sign-on and unified financial tracking



Analytic Center as a Framework





AC Concept of Operations

- Principles
 - Each researcher defines their own process
 - Community accepted tools allow re-use of others' work
 - Unique tools and data should be easily integrated
 - Collect publication materials as you go, in background
- Notional Process
 - PI defines hypothesis and process
 - Select framework to be used and workflow tool
 - Shop for tools, data and compute resources
 - Assemble components and verify integrity
 - Perform analysis, develop models, visualize, etc.
 - Draw conclusions and perform internal reviews
 - Enter publication process



AIST Experiments with Analytic Centers

- Land Use and Land Change
 - NEX (NASA Ames) Rama Neimani
- Physical Oceanography
 - OceanWorks (JPL) Thomas Huang
- Tropical Cyclones and Hurricanes
 - TCIS (JPL) Svetla Hristova-Veleva
- Climate
 - Climate Model Diagnostic Analyzer (CMDA), Seungwon Lee (JPL)
 - Climate Workbench (UAH/MSFC) Manil Maskey, Chris Lynnes (GSFC)
- Communities being discussed for further experiments
 - Biodiversity
 - Hydrology
 - Atmospheric Composition
 - Cryosphere



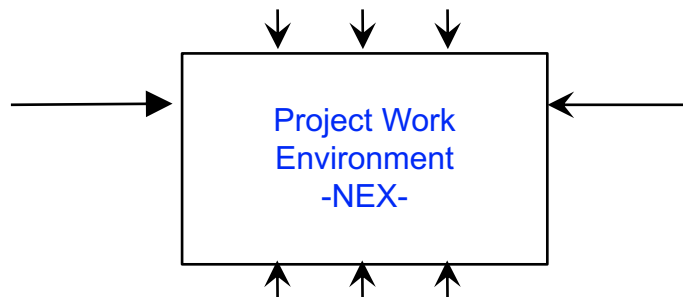
NASA Earth Exchange (NEX) as an Analytic Center

Land Change/Use Community

- Project Definition
- Plan for Investigation

Data

- Built into website
- Dataset Sources
 - Landsat
 - Sentinel 1A
 - Modis
 - ASTER
 - TRMM
 - AVHRR
- Climate Datasets
- Land cover
- Digital Elevation Map
- STATSGO Soils
- USDA Aerial NAIP
- And others



Tools

- Models
 - Tops
 - Biome-BGC
 - LPJ Dynamic Global Model
- Sandbox for small scale experiments
- Analysis
 - R and python based tools
 - Matlab VIIRS HDF5 swath conversion
- Workflow: Jupyter Notebook

Computational Infrastructure

Storage

- Data Containers
- Thematic model
- Metadata/Ontology
- Resulting Products
- Published data
- Provenance

Computing

- NASA Advanced Supercomputing (NAS)
- Amazon Web Services (AWS) Public Cloud
- Ames Quantum Computer (D-wave)

POC: rama.nemani@nasa.gov

<https://nex.nasa.gov/nex>



A Strategy for Introduction

- Focus on needs of a particular Science community
 - Data, models and tools they are comfortable with
 - Computational resources adequate to perform their investigation
 - Concept of Operations must support their work, not force a fit
 - Identify needs, lessons learned, community perceptions
 - Are there common processes that could make it easier
- NASA's Earth Science needs **much** more exposure to Machine Learning
 - Researchers and Managers need to develop confidence in the tools
 - Outreach to the Science community
- Must be an architecture with modularity and interoperability
- Must support security
- Increase access to non-NASA data sources as well as NASA
- AIST will experiment with basic concepts as well as components

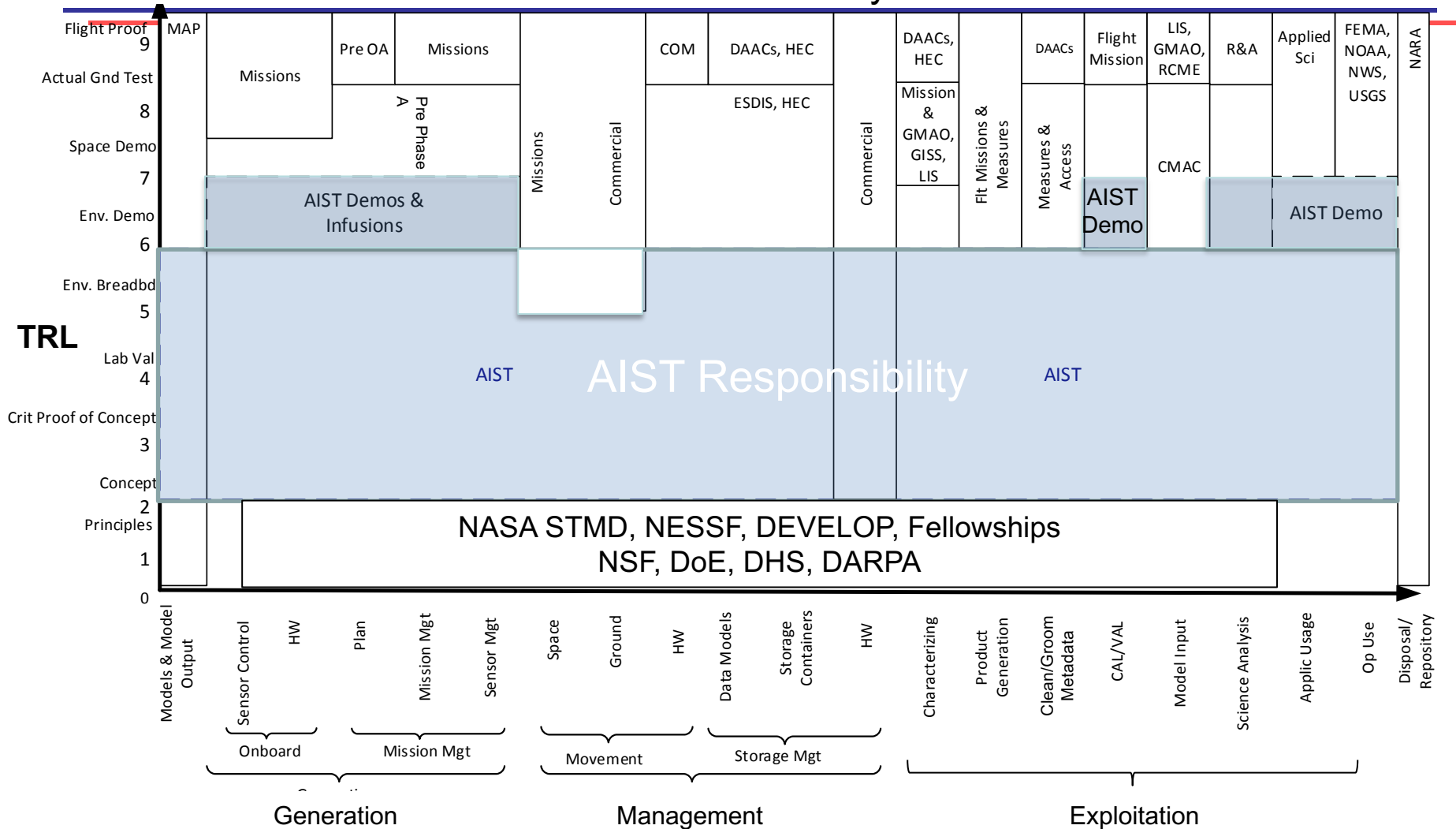


BACK UP



A View of AIST

TRL v. Data Life Cycle



Data Life Cycle

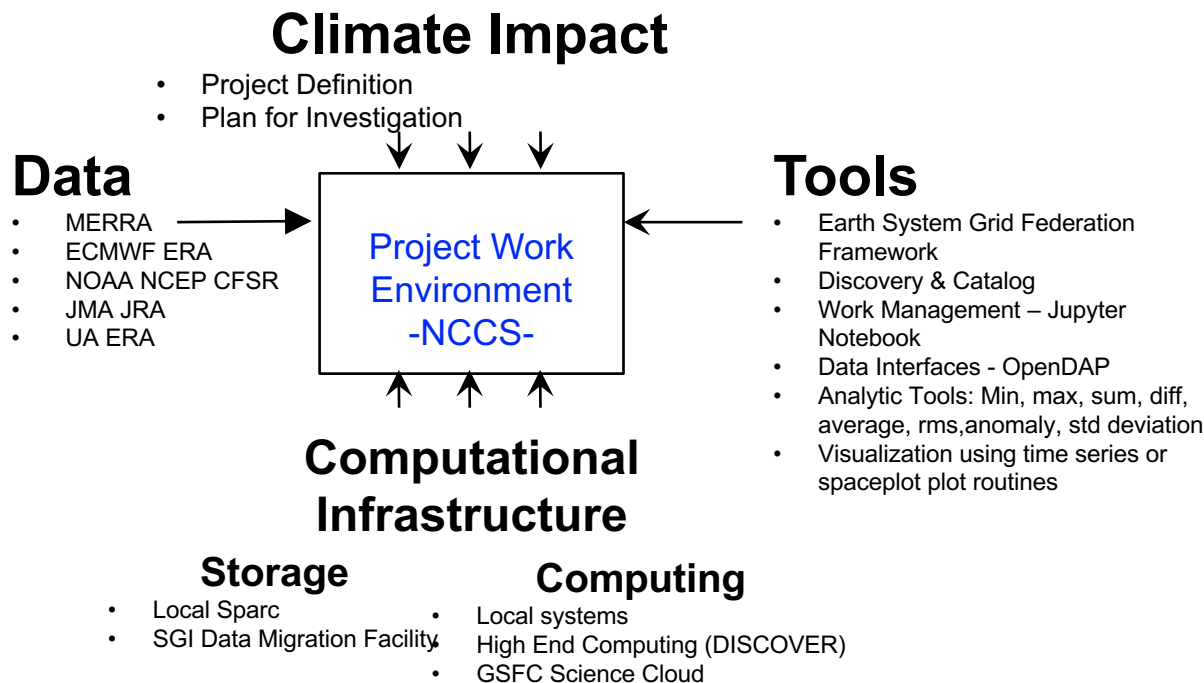


AC ConOps Detailed Sample

- PI defines the investigation to be done
 - Types of data, analysis steps, process
- Select workflow tool to manage process
 - Such as Jupyter Notebook
- Shop for data, tools and computing resources needed
 - Parameter search of known data repositories
 - Find more like...
 - Identify data from local or unpublished sources
 - Create metadata to make it usable by others, including peer review
- Integrate local data into local storage – compatibility mode
- Retrieve remote data into local storage
 - Storage method is harmonized with known tools
 - Verifies accuracy of copy compared to repository
 - Perform incremental copy/analysis if storage limited
- Integrate local tools into work flow to interact with data and storage effectively
 - Document the tool for future reference and publication
- Massage data as required (geolocation, gridding, characteristics)
- Step through analytic steps with a limited subset of data using a limited set of processing resources
 - Visualize or analyze results until satisfied with process
- Scale up processing of entire data set
- Review results and draw conclusions
- Review AC publication package
 - Source code, tools, local data, metadata, etc.
- Submit for publication



Earth Data Analytics System (EDAS) as an Analytic Center



POC: laura.carriere@nasa.gov

<https://www.nccs.nasa.gov/services/Analytics>



Tragedy,

- Large investments by DoD and IC in AI and Machine Learning have, largely, not been leveraged by NASA Earth Science
- Lack of credibility in ML by Earth Science researchers
 - Why does ML generate bogus conclusions?
 - Failure to articulate assumptions & constraints and exceeding them
 - Overtraining
 - Drift over time
 - Failure to recognize when conditions change
 - Earth Scientists have trouble understanding the technology
 - Lack of demonstration of value to motivate them
 - Lack of training in the tools and algorithms
- Partnerships among Machine Learning and Earth Science communities are slow to take root
 - Communications
 - Opportunity
 - Trust
- Data grooming takes huge amounts of time



Myths

- You can do a problem and set it aside for others to use.
 - There is no fire and forget in Earth Science
- Open source is always better than commercial software
 - GIS is a good counter-example
 - Communicating with other Agencies requires arc-GIS capability
- Data must be supplied in same form as collected
 - Stewardship is important, but not useful for analysis
 - Must be able to trace back to authoritative data
- Cannot re-use (trust) anyone else's work
 - Community acceptance and jurying could create common capabilities
 - Fix the data once and others could use it if they believe it is fixed



Analytic Center Characteristics

- Seamless integration
- Comprehensive catalog
 - Clear applicability (or not)
 - Expert system as an operator aid
- Help in using them
 - YouTube, moodle